

A Market Basket Analysis Product Recommendation System for Online Retailers

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Abstract

Using predictive analytics, we developed a market basket analysis model that will predict future purchases given historical purchasing patterns. Overall, this will help consumers become more aware of products available and contribute directly to the vendor's performance. Our study provides a quantitative and associative design to this problem, where we show the business benefits of using the model to provide intelligent recommendations.

Introduction

In order to predict what products a customer will purchase together in the future, we developed a market basket analysis model given their historical purchasing patterns. For traditional in-store shopping retailers such as Kroger and Meijer, this research is important for developing intelligent product recommendations that predict a consumer's possible buying behaviors. This model will greatly benefit consumers and will increase retailer performance.



Picture source: http://www.salemmarafi.com/code/market-basket-analysis-with-r/

Research questions:

- 1. Based on purchases history, what products will a customer likely purchase together in the future?
- 2. How does the usage of this recommendation system affect both the consumer and the retailer?

Literature Review

In recent studies, a Market Basket Analysis finds a common association between different items that are purchased while further analyzing their data using the Apriori Algorithm. Generally, supermarket data consists of products that are typically purchased, but does not include items that were repurchased.

Study	Association Rules	Apriori Algorithm	Supermarket	Reordering
(Ting, 2010)	\checkmark			
(Abubakar, 2015)	\checkmark	\checkmark	~	
(Abdulsalam, 2014)	\checkmark	\checkmark	~	
(Musungwini, 2014)	\checkmark		~	
Our Study	\checkmark	\checkmark	\checkmark	\checkmark

Table 1: Literature Review Summary

Our study is novel because we are using similar methods, but incorporate the reordering factor to better analyze the associations.

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Results Frequent Items support: 0.01 Bag of Organic Bananas Organic Strawberries Organic Baby Spinach This chart shows the Large Lemon frequency of single items Limes Organic Hass Avocado Strawberries purchased by new Organic Avocado Organic Blueberries Organic Garlic Organic Yĕllow Onion customers. Organic Zucchini Organic Raspberries Yellow Onions Cucumber Kirby anic Grape Tomatoes We found that Bananas eedless Red Grapes Organic Baby Carrots Organíc Lemon Carrots were the most frequent Extra Virgin Olive Oil Organic Cilantro Organic Fuji Apple item purchased. Asparagus 4000 Figure 5: Single Items Purchased- Frequency Frequent Itemsets support: 0.001 This chart shows the {Bag of Organic Bananas, Organic Strawberries} {Banana,Organic Baby Spinach} frequency of items bought {Bag of Organic Bananas, Organic Baby Spinach} Banana,Strawberries} Organic Avocado, Organic Baby Spinach} together and the {Limes,Organic Cilantro} {Organic Garlic.Organic Yellow Onion} associations we found {Banana,Large Lemon} {Banana,Limes} between those items {Organic Avocado, Organic Strawberries} {Banana,Cucumber Kirby} {Bag of Organic Bananas, Organic Raspberries} {Limes,Organic Garlic} {Limes, Organic Baby Spinach We found {Bananas, {Bag of Organic Bananas, Organic Blueberries} {Banana, Honeycrisp Apple} Organic Strawberries} to {Limes,Organic Hass Avocado} {Limes, Organic Strawberries} {Banana,Organic Blueberries} be the highest association {Organic Garlic, Organic Ginger Root} **Figure 6: Associations Found- Frequency** Scatter plot for 173971 rules This Lift Chart shows that 8000 there are some rules with

0.0003 0.0004 0.0001 0.0002 suppor

Figure 7: Lift Chart

a heavy lift indicating a strong association between the items.

Conclusions

This problem is important because companies are relying more on data centered analysis of what their customers will buy now and in the future.

I. Based on purchase history, we found that people tend to purchase fresh fruit, especially when they are returning customers.

2. This model helps customers increase efficiency while shopping and helps retailers better place their items and promotions and potentially increase store sales and revenues.

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